

Application of Information Theory, Lecture 11

Pseudo-Entropy and Pseudorandom Generators

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January 6, 2015

Part I

Motivation

Encryption schemes

Definition 1

A pair of algorithms (E, D) is (perfectly correct) encryption scheme, if for any $k \in \{0, 1\}^n$ and $m \in \{0, 1\}^\ell$, it holds that $D(k, E(k, m)) = m$

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- ▶ Statistical security? HW. Computational security?

Part II

Statistical Vs. Computational distance

Distributions and statistical distance

Let \mathcal{P} and \mathcal{Q} be two distributions over a finite set \mathcal{U} . Their **statistical distance** (also known as, variation distance) is defined as

$$\text{SD}(\mathcal{P}, \mathcal{Q}) := \frac{1}{2} \sum_{x \in \mathcal{U}} |\mathcal{P}(x) - \mathcal{Q}(x)| = \max_{S \subseteq \mathcal{U}} (\mathcal{P}(S) - \mathcal{Q}(S))$$

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Claim 2

For any pair of (finite) distributions \mathcal{P} and \mathcal{Q} , it holds that

$$\text{SD}(\mathcal{P}, \mathcal{Q}) = \max_{\mathcal{D}} \{ \Delta^{\mathcal{D}}(\mathcal{P}, \mathcal{Q}) := \Pr_{x \leftarrow \mathcal{P}} [\mathcal{D}(x) = 1] - \Pr_{x \leftarrow \mathcal{Q}} [\mathcal{D}(x) = 1] \},$$

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Let $\mathcal{P}, \mathcal{Q}, \mathcal{R}$ be finite distributions, then

Triangle inequality: $\text{SD}(\mathcal{P}, \mathcal{R}) \leq \text{SD}(\mathcal{P}, \mathcal{Q}) + \text{SD}(\mathcal{Q}, \mathcal{R})$

Repeated sampling: $\text{SD}(\mathcal{P}^2 = (\mathcal{P}, \mathcal{P}), \mathcal{Q}^2 = (\mathcal{Q}, \mathcal{Q})) \leq 2 \cdot \text{SD}(\mathcal{P}, \mathcal{Q})$

Section 1

Computational Indistinguishability

Computational indistinguishability

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\mathcal{P} and \mathcal{Q} are (s, ϵ) -indistinguishable, if $\Delta_{\mathcal{P}, \mathcal{Q}}^D \leq \epsilon$, for any s -size D .

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- ▶ We sometimes think of $s = n^{\omega(1)}$ and $\epsilon = 1/s$, where n is the “security parameter”
- ▶ Can it be different from the statistical case?
- ▶ Unless said otherwise, distributions are over $\{0, 1\}^n$

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Question 4

Assume \mathcal{P} and \mathcal{Q} are (s, ϵ) -indistinguishable, what about \mathcal{P}^2 and \mathcal{Q}^2 ?

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- ▶ So either $\Delta^D_{(\mathcal{P}^2, (\mathcal{P}, \mathcal{Q}))} \geq \varepsilon'/2$, or $\Delta^D_{((\mathcal{P}, \mathcal{Q}), \mathcal{Q}^2)} \geq \varepsilon'/2$

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- ▶ Hence, $\varepsilon' < 2\varepsilon$ implies $s' \geq s - 2n$.

Repeated sampling cont.

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Claim 5

Assume \mathcal{P} and \mathcal{Q} are (s, ε) -indistinguishable, then \mathcal{P}^k and \mathcal{Q}^k are $(s - kn, k\varepsilon)$ -indistinguishable.

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Claim 5

Assume \mathcal{P} and \mathcal{Q} are (s, ϵ) -indistinguishable, then \mathcal{P}^k and \mathcal{Q}^k are $(s - kn, k\epsilon)$ -indistinguishable.

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- ▶ Thus, $\varepsilon' \leq k\varepsilon$ implies $s' > s - kn$
- ▶ When considering bounded **time** algorithms, things behaves very differently!

Part III

Pseudorandom Generators

Pseudorandom generator

Definition 6 (pseudorandom distributions)

A distribution \mathcal{P} over $\{0, 1\}^n$ is (s, ϵ) -pseudorandom, if it is (s, ϵ) -indistinguishable from U_n .

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A poly-time computable function $g: \{0, 1\}^n \mapsto \{0, 1\}^{\ell(n)}$ is a (s, ε) -pseudorandom generator, if for any $n \in \mathbb{N}$

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- ▶ Do such generators exist?
- ▶ Applications?

Section 2

Pseudorandom generators (PRGs) from One-Way Permutations (OWPs)

OWP to PRG

Claim 8

Let $f: \{0, 1\}^n \mapsto \{0, 1\}^n$ be a poly-time permutation and let $b: \{0, 1\}^n \mapsto \{0, 1\}$ be a poly-time (s, ε) -hardcore predicate of f , then $g(x) = (f(x), b(x))$ is a $(s - O(n), \varepsilon)$ -PRG.

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- ▶ Proof: Let D be an s' -size algorithm with $\Delta_{g(U_n), U_{n+1}}^D = \varepsilon'$, we will show \exists $(s' + O(n))$ -size P with $\Pr[P(f(U_n)) = b(U_n)] = \frac{1}{2} + \varepsilon'$.

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$$\delta = \Pr[D(f(U_n), U_1) = 1]$$

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- ▶ Hence, $\Pr[D(f(U_n), \overline{b(U_n)}) = 1] = \delta - \varepsilon'$

OWP to PRG cont.

- ▶ $\Pr[D(f(U_n), b(U_n)) = 1] = \delta + \varepsilon'$
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Algorithm 9 (P)

Input: $y \in \{0, 1\}^n$

1. Flip a random coin $c \leftarrow \{0, 1\}$.
2. If $D(y, c) = 1$ output c , otherwise, output \bar{c} .

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Part IV

PRG from Regular OWF

Computational notions of entropy

Definition 10

X has (s, ε) -pseudoentropy at least k , if \exists rv Y with $H(Y) \geq k$ and $\Delta^D(X, Y) \leq \varepsilon$ for any s -size D .

(s, ε) -pseudo min/Reiny -entropy are analogously defined.

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► Example

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- ▶ Example
- ▶ Repeated sampling

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- ▶ Example
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- ▶ Example
- ▶ Repeated sampling
- ▶ Non-monotonicity
- ▶ Ensembles
- ▶ In the following we will simply write (s, ϵ) -entropy, etc

High entropy OWF from regular OWF

Claim 11

Let $f: \{0, 1\}^n \mapsto \{0, 1\}^n$ be a 2^k -regular (s, ε) -one-way, let $\mathcal{H} = \{h: \{0, 1\}^n \mapsto \{0, 1\}^{m=k+\lceil \log n \rceil}\}$ be 2-universal family, and let $g(h, x) = (g(x), h, h(x))$. Then

1. $H_2(g(U_n, H)) \geq 2n - \frac{1}{n}$, for $H \leftarrow \mathcal{H}$.
2. g is $(\Theta(s\varepsilon^2/n), 2\varepsilon)$ -one-way.

- ▶ k and m and \mathcal{H} are parameterized by of n
- ▶ We assume $\log |\mathcal{H}| = n$ and $s \geq n$

g has high Renyi entropy

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$$\begin{aligned}\text{CP}(g(U_n, H)) &:= \Pr_{w, w' \leftarrow \{0,1\}^n \times \mathcal{H}} [g(w) = g(w')] \\ &= \Pr_{h, h' \leftarrow \mathcal{H}} [h = h'] \cdot \Pr_{(x, x') \leftarrow (\{0,1\}^n)^2} [f(x) = f(x')] \\ &\quad \cdot \Pr_{h \leftarrow \mathcal{H}; (x, x') \leftarrow (\{0,1\}^n)^2} [h(x) = h(x') \mid f(x) = f(x')]\end{aligned}$$

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Hence, $H_2(g(U_n, H)) \geq H_2(\mathcal{H}) + H_2(U_n) + \log \frac{1}{1 + \frac{1}{n}} \geq n + n - \frac{1}{n}$.

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Algorithm 12 (B)

Input: $y \in \{0, 1\}^n$.

Return $D(y, h, z)$, for $h \leftarrow \mathcal{H}$ and $z \leftarrow \{0, 1\}^\ell$.

Algorithm 13 (D)

Input: $y \in \{0, 1\}^n$, $h \in \mathcal{H}$ and $z_1 \in \{0, 1\}^\ell$.

For all $z_2 \in \{0, 1\}^{m-\ell}$:

1. Let $(x, h) = A(y, h, z_1 \circ z_2)$.
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- ▶ $\Pr_{x \leftarrow \{0,1\}^n; h \leftarrow \mathcal{H}} [D(f(x), h, h(x)_{1,\dots,\ell}) \in f^{-1}(f(x))] \geq \varepsilon'$

g is one-way, cont.

We saw that

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By the leftover hash lemma

$$\text{SD}((f(x), h, h(x)_1, \dots, h(x)_\ell)_{x \leftarrow \{0,1\}, h \leftarrow \mathcal{H}}, (f(x), h, U_\ell)_{x \leftarrow \{0,1\}, h \leftarrow \mathcal{H}}) \leq \varepsilon'/2 \quad (2)$$

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Hence,

$$\Pr_{x \leftarrow \{0,1\}^n} [B(f(x)) \in f^{-1}(f(x))] \geq \varepsilon' - \varepsilon'/2 = \varepsilon'/2.$$

The generator

Claim 14

Let $g: \{0, 1\}^n \mapsto \{0, 1\}^m$ be a function with $H_2(f(U_n)) \geq n - \frac{1}{2}$, and let b be (s, ϵ) -hardcore predicate for g . Then $v(U_n) = (g(U_n), b(U_n))$ has (s, ϵ) -Renyi-entropy $n + \frac{1}{2}$.

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The generator cont.

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Claim 16

Let $\mathcal{H}: \{0, 1\}^{n^2+n} \mapsto \{0, 1\}^{n^2+n/4}$ be an 2-universal family and let $G: \{0, 1\}^n \times \mathcal{H}$ defined by $G(x_1, \dots, x_n, h) = (h, h(v^n(x_1, \dots, x_n)))$. Then $G(H, U_n^n)$ is $(s - ns_v - s_{\mathcal{H}}, n\varepsilon + 2^{-n/4})$ indistinguishable from $(H, U_{n^2+n/4})$, for $H \leftarrow \mathcal{H}$ and $s_{\mathcal{H}}$ being the size of sampling and evaluating algorithm for \mathcal{H} .

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Corollary 17

If f and b and \mathcal{H} (?) are poly-time computable, then G is a $(s - n^2 - s_{\mathcal{H}}, n\varepsilon + 2^{-n/4})$ -PRG.

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If f and b and \mathcal{H} (?) are poly-time computable, then G is a $(s - n^2 - s_{\mathcal{H}}, n\varepsilon + 2^{-n/4})$ -PRG.

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- By the leftover hash lemma $SD((H, H(Z^n)), (H, U_{n^2+n/4})) \leq 2^{-n/4}$

The generator cont.

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- ▶ Hence, $\exists (s' + s_{\mathcal{H}})$ -size algorithm that distinguishes $G(U_n^n, H)$ from $(H, H(Z^n))$ with advantage ε'
- ▶ Hence $s' \leq s - n^2 - s_{\mathcal{H}} \implies \varepsilon' \leq n\varepsilon$.

The generator cont.

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Let $\mathcal{H}: \{0, 1\}^{n^2+n} \mapsto \{0, 1\}^{n^2+n/4}$ be an 2-universal family and let $G: \{0, 1\}^n \times \mathcal{H}$ defined by $G(x_1, \dots, x_n, h) = (h, h(v^n(x_1, \dots, x_n)))$. Then $G(H, U_n^n)$ is $(s - ns_v - s_{\mathcal{H}}, n\varepsilon + 2^{-n/4})$ indistinguishable from $(H, U_{n^2+n/4})$, for $H \leftarrow \mathcal{H}$ and $s_{\mathcal{H}}$ being the size of sampling and evaluating algorithm for \mathcal{H} .

Corollary 17

If f and b and \mathcal{H} (?) are poly-time computable, then G is a $(s - n^2 - s_{\mathcal{H}}, n\varepsilon + 2^{-n/4})$ -PRG.

Proof: (of claim)

- ▶ By the leftover hash lemma $\text{SD}((H, H(Z^n)), (H, U_{n^2+n/4})) \leq 2^{-n/4}$
- ▶ Let D be an s' -size algorithm that distinguishes $G(U_n^n, H)$ from $(H, U_{n^2+n/4})$ with advantage $\varepsilon' + 2^{-n/4}$
- ▶ Hence, $\exists (s' + s_{\mathcal{H}})$ -size algorithm that distinguishes $G(U_n^n, H)$ from $(H, H(Z^n))$ with advantage ε'
- ▶ Hence $s' \leq s - n^2 - s_{\mathcal{H}} \implies \varepsilon' \leq n\varepsilon$.

The generator cont.

Claim 16

Let $\mathcal{H}: \{0, 1\}^{n^2+n} \mapsto \{0, 1\}^{n^2+n/4}$ be an 2-universal family and let $G: \{0, 1\}^n \times \mathcal{H}$ defined by $G(x_1, \dots, x_n, h) = (h, h(v^n(x_1, \dots, x_n)))$. Then $G(H, U_n^n)$ is $(s - ns_v - s_{\mathcal{H}}, n\varepsilon + 2^{-n/4})$ indistinguishable from $(H, U_{n^2+n/4})$, for $H \leftarrow \mathcal{H}$ and $s_{\mathcal{H}}$ being the size of sampling and evaluating algorithm for \mathcal{H} .

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Remarks

- ▶ PRG “length extension”

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